STAT C206B: Topics in Stochastic Processes

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1 Lecture 1: Backround material

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1.1 Ferguson distributions / Dirichlet processes

Definition 1 (Gamma distribution)

Random variable X supported on $(0, \infty)$ has **Gamma distribution** with shape $\alpha > 0$ and inverse scale / rate $\beta > 0$, written $X \sim \text{Gamma}(\alpha, \beta)$ if it has density

$$f_X(t) = \mathbb{1}\{t \in (0, \infty)\} \frac{\beta^{\alpha} t^{\alpha - 1} e^{-\beta t}}{\Gamma(\alpha)}$$
(1)

where $\Gamma(t) = \int_0^\infty u^{t-1} e^{-u} du$ is the Gamma function defined for all $\Re t > 0$ and analytically continued to $\mathbb{C} \setminus \{n \in \mathbb{Z} : n < 0\}$

Proposition 2 (Gamma closed under summation)

If $Y \sim \text{Gamma}(\alpha, \beta)$ and $Z \sim \text{Gamma}(\gamma, \beta)$ are independent, then $Y + Z \sim \Gamma(\alpha + \gamma, \beta)$.

{prop:gammaclosed-sum}

Proof.

$$\begin{split} f_{Y+Z}(t) &= \int_0^t f_Y(u) f_Z(t-u) du \\ &= \frac{1}{\Gamma(\alpha) \Gamma(\gamma)} \beta^{\alpha+\gamma} e^{-\beta t} \int_0^t u^{\alpha-1} (t-u)^{\gamma-1} du \\ &= \frac{1}{\Gamma(\alpha) \Gamma(\gamma)} \beta^{\alpha+\gamma} e^{-\beta t} \int_0^1 (tv)^{\alpha-1} (t-(tv))^{\gamma-1} t dv \\ &= \frac{1}{\Gamma(\alpha) \Gamma(\gamma)} \beta^{\alpha+\gamma} e^{-\beta t} t^{\alpha+\gamma-1} B(\alpha, \gamma) \end{split}$$

where $B(x,y)=\int_0^1 t^{x-1}(1-t)^{y-1}dt=\frac{\Gamma(x)\Gamma(y)}{\Gamma(x+y)}$ is the beta function

A closely related distribution obtained from concatenating Gamma random variables into a vector and then normalizing the sum to 1 is the Dirichlet distribution.

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Definition 3 (Dirichlet distribution)

Let $\alpha \in (0, \infty)^K$. Random (probability) vector X taking values on the K-1-dimensional probability simplex $\Delta^{K-1} = \{ \boldsymbol{x} \in [0, 1]^K : \sum_i x_i = 1 \}$ has **Dirichlet distribution** of order K and concentration parameters $\boldsymbol{\alpha}$, denoted $X \sim \text{Dir}(\boldsymbol{\alpha})$, if it has density

$$f_X(\boldsymbol{x}) = \mathbb{1}\{\boldsymbol{x} \in \Delta\} \underbrace{\frac{\Gamma\left(\sum_{i=1}^K \alpha_i\right)}{\prod_{i=1}^K \Gamma(\alpha_i)}}_{=:B(\boldsymbol{\alpha})^{-1}} \prod_{i=1}^K x_i^{\alpha_i - 1}$$

Proposition 4 (Constructing Dirichlet from Gammas)

{prop:dirich let-from-gam ma}

Let $X_1, \ldots, \overline{X_n}$ be independent $\operatorname{Gamma}(\alpha_i, \beta)$ distributed, $S_n = \sum_{i=1}^n X_i$. Then $(V_i)_i = (X_i/S_n)_i \sim \operatorname{Dir}(\boldsymbol{\alpha})$.

Proof. $S_n \sim \Gamma(\sum_{i=1}^n \alpha_i, \beta)$ by Proposition 2 and for $v \in \Delta^{n-1}$, we have

$$f_{V}(\boldsymbol{v}) = \int_{0}^{\infty} f_{X}\left(sv_{1}, \dots, sv_{n-1}, sv_{n}\right) f_{S_{n}}(s) ds$$

$$= \int_{0}^{\infty} e^{-\sum_{i=1}^{n} sv_{i}} \left(\prod_{i=1}^{n} \frac{(sv_{i})^{\alpha_{i}-1}}{\Gamma(\alpha_{i})} \right) \frac{s^{\sum_{i}^{n} \alpha_{i}-1} e^{-s}}{\Gamma(\sum_{i}^{n} \alpha_{i})} ds$$

$$= \frac{1}{\prod_{1}^{n} \Gamma(\alpha_{i})} \prod_{i=1}^{n} v_{i}^{\alpha_{i}-1} \int_{0}^{\infty} e^{-s \sum_{1}^{n} \sigma_{i}^{*}} \int_{s}^{t} \sum_{1}^{n} \alpha_{i}^{*} e^{-t} ds$$

$$= \frac{\Gamma(\sum_{i=1}^{n} \alpha_{i})}{\prod_{1}^{n} \Gamma(\alpha_{i})} \prod_{i=1}^{n} v_{i}^{\alpha_{i}-1}$$

Similar to Proposition 2 (Gamma closed under summation), where adding two Gammas yielded another Gamma where the parameters were added, Dirichlet distributions enjoy a similar kind of closure: "clumping" coordinate axes together (described below) yields another Dirichlet distribution where the parameters of the clumped axes are summed together.

Proposition 5 (Dirichlet clumping property)

Suppose $X \sim \text{Dir}(\alpha_1, \dots, \alpha_n)$. For any $r \leq n$, let $V_i = X_i$ for $i \in [r]$ and let $V_{r+1} = \sum_{j=r+1}^n X_j$. Then $V \sim \text{Dir}(\alpha_1, \dots, \alpha_r, \sum_{j=r+1}^n \alpha_j)$.

Proof. By induction, it suffices to show this for r = n - 2. Notice

$$f(v_1, \dots, v_r, s) = B(\boldsymbol{\alpha})^{-1} \left(\prod_{i=1}^{n-1} v_i^{\alpha_i - 1} \right) \int \mathbbm{1} \left\{ x_{n-1} + x_n = s \right\} x_{n-1}^{\alpha_{n-1} - 1} x_n^{\alpha_n - 1} dx_n$$

$$= B(\boldsymbol{\alpha})^{-1} \left(\prod_{i=1}^{n-1} v_i^{\alpha_i - 1} \right) \int_0^s u^{\alpha_{n-1} - 1} (s - u)^{\alpha_n - 1} du$$

$$= B(\boldsymbol{\alpha})^{-1} \left(\prod_{i=1}^{n-1} v_i^{\alpha_i - 1} \right) s^{\alpha_{n-1} + \alpha_n - 1} B(\alpha_{n-1}, \alpha_n)$$
Since $\frac{B(\alpha_{n-1}, \alpha_n)}{B(\boldsymbol{\alpha})} = \frac{\Gamma(\sum_{i=1}^n \alpha_i)}{\Gamma(\alpha_{n-1} + \alpha_n) \prod_{i=1}^{n-2} \Gamma(\alpha_i)}$, we are done.

Iterating this result over coordinate axes other than the last n-r, we see that "clumping together" entries in a Dirichlet random vector yields another Dirichlet random vector with parameters also "clumped together." Concretely, for any mapping $\phi: [n+1] \to [m+1]$ if $U_j = \sum_{\phi(i)=j} V_i$ then U has Dirichlet distribution with parameters $\gamma_j = \sum_{\phi(i)=j} \alpha_i$.

Generalizing this clumping property is the motivation for $Ferguson\ Distributions$ (Ferguson, 1973).

Definition 6 (Ferguson / Dirichlet process distribution)

Let μ be a finite positive Borel measure on complete separable metric space E. A random probability measure μ^* on E (i.e. a stochastic process indexed by a σ -algebra on E) has **Ferguson distribution** with parameter μ if for every finite partition $(B_i)_{i \in [r]}$ of E the random vector

$$(\mu^*(B_i))_{i\in[r]} \sim \operatorname{Dir}(\mu(B_1), \dots, \mu(B_r))$$

Lemma 7 (Preservation of Ferguson under pushforward)

Let μ^* be Ferguson with parameter μ and $\phi: E \to F$ measurable. Then the pushforward $\mu^* \circ \phi^{-1}$ is a random probability measure on F that has Ferguson distribution with parameter $\mu \circ \phi^{-1}$.

Proof. For $(B_i)_{i \in [r]}$ a finite partition of F, $(\phi^{-1}(B_i))_i$ is a finite partition of E. Since μ^* is Ferguson

$$(\mu^*(\phi^{-1}(B_i)))_i \sim \text{Dir}((\mu(\phi^{-1}(B_i)))_i)$$

Hence $\mu^* \circ \phi^{-1}$ is Ferguson with parameter $\mu \circ \phi^{-1}$.

Next, we turn to an important class of a Ferguson distributions arising from generalizing the Pólya urn.

Definition 8 (Polya sequence)

A sequence $(X_n)_{n\in\mathbb{N}}$ with values in E is a **Polya sequence with parameter** μ if for all $B\subset E$.

$$\Pr[X_1 \in B] = \mu(B)/\mu(E)$$

$$\Pr[X_{n+1} \in B \mid X_1, \dots, X_n] = \mu_n(B)/\mu_n(E)$$

where $\mu_n = \mu + \sum_{i=1}^n \delta_{X_i}$.

Remark 9. When E is finite (e.g. a set of colors for the balls), (X_n) represents the result of successive draws from an urn with initially $\mu(x)$ balls of color $x \in E$ and after each draw a ball of the same color as the one drawn is added back to give an urn with color distribution $\mu_{n+1}(x)$.

blackwell11973 Blackwell and MacQueen (1973) gives the following result connecting Pólya sequences and Ferguson distributions.

Theorem 10 (Polya Urn Schemes)

Let (X_n) be a Polya sequence with parameter μ . Then:

- 1. $m_n = \mu_n/\mu_n(E)$ converges almost surely to a limiting discrete measure μ^*
- 2. μ^* has Ferguson distribution with parameter μ
- 3. Given μ^* , $(X_i)_{i>1}$ are independent with distribution μ^*

Proof. First consider E finite and let μ^* and $\{X_i\}$ be random variables whose joint distribution satisfies (2.) and (3.). Let π_n be empirical distribution of $(X_i)_{i \in [n]}$. $X_i \stackrel{\text{iid}}{\sim} \mu^*$, so by SLLN $\pi_n \stackrel{as}{\to} \mu^*$ and since

$$m_n = \frac{\mu + n\pi_n}{\mu(E) + n} \tag{2}$$

(1.) follows.

It remains to show (X_n) is a Polya sequence with parameter μ , i.e.

$$\Pr[A] = \prod_{x} \mu(x)^{[n(x)]} / \mu(E)^{[n]} \tag{3} \quad \{\{\text{eq:polya-s eq-meas}\}\}$$

where $A = \{X_i = x_i\}_i$, $n(x) = \#\{i : x_i = x\}$, and the rising factorial $a^{[k]} = a(a+1)\cdots(a+k-1)$. Notice

$$\Pr[A] = \mathbb{E}\left[\Pr[A \mid \mu^*]\right] = \mathbb{E}\left[\prod_x \mu^*(x)^{n(x)}\right]$$
(4)

Since μ^* is Ferguson, viewing $E = \bigsqcup_{x \in E} \{x\}$ as a partition we have $(\mu^*(x))_{x \in E} \sim \text{Dir}((\mu(x))_{x \in E})$ so the RHS is the $(n(x))_{x\in E}$ moment of the Dirichlet distribution, which is equal to

$$\mathbb{E}\left[\prod_{x} \mu^{*}(x)^{n(x)}\right] = \frac{\Gamma(\mu(E))}{\Gamma(\mu(E) + n)} \prod_{x} \frac{\Gamma(\mu(x) + n(x))}{\Gamma(\mu(x))} = \frac{1}{\mu(E)^{[n]}} \prod_{x} \mu(x)^{[n(x)]}$$
(5)

as required by Eq. (3).

General E follows from approximation argument.

We leave the discreteness part of (1.) as an exercise, noting that similar to how Dirichlets can be defined as a set of independent Gammas normalized by their sum (Proposition 4 (Constructing Dirichlet from Gammas)) we would expect the Dirichlet process / Ferguson random measures to be definable as a gamma process with independent "increments" divided by their sum.

Exercise 11. Prove every Ferguson random measure is discrete. (Hint: argue using moments).

Remark 12. If (X_n) a Polya sequence, then it is a mixture of iid sequences, and is exchangeable i.e. $(X_i) \stackrel{d}{=} (X_{\sigma(i)})$ (see Eq. (3))

Construction of Haar Measure

For a finite group G, the measure $\mu(g) = \frac{1}{\#G}$ is left and right translation invariant i.e. $\mu(gA) = \mu(A) = \mu(A)$ $\mu(Ag)$.

In fact, all compact groups have unique translation invariant measure called the Haar measure. For example $Z_{ij} \stackrel{\text{iid}}{\sim} N(0,1)$ for $i,j \in [n]$ and X the Gram-Schmidt orthonormalization of the rows of Z. Then $XU \stackrel{d}{=} UX$ for all $U \in O(n)$, so X has Haar measure on compact group O(n).

Definition 13

A topological vector space (TVS) is a vector space equipped with a topology such that vector space operations are jointly continuous.

Example 14

 \mathbb{R}^n with standard topology, any Banach space.

Definition 15

A family $\mathfrak G$ of linear transformations on TVS $\mathfrak X$ is equicontinuous on subset $K \subset \mathfrak X$ if for every neighborhood V of the origin, there exists a neighborhood U of the origin such that

$$\forall k_1, k_2 \in K : k_1 - k_2 \in U \Rightarrow \mathfrak{G}(k_1 - k_2) \subset V \tag{6}$$

That is, $T(k_1 - k_2) \in V$ for all $T \in \mathfrak{G}$.

Definition 16

i}

A locally convex topological vector space (LCTVS) is a TVS with a local base of absolutely convex absorbing sets at the origin.

To construct Haar measure for any compact group, we will need a fix point theorem due to Kakutani.

Theorem 17 (Kakutani Fix Point Theorem)

{thm:kakutan

K compact convex subset of LCTVS \mathfrak{X} , \mathfrak{G} group of linear transforms equicontinuous on K and such that $\mathfrak{G}(K) \subset K$, then $\exists p \in K$ such that

$$\mathfrak{G}(p) = \{p\} \tag{7}$$

• By Zorn's lemma applied to chains $(K_{\lambda})_{\lambda}$ (note $K_a \subset K_b$ for a < b), \exists minimal $K_1 \subset K$ such that $K_1 \neq \emptyset$ and $\mathfrak{G}(K_1) \subset K_1$.

- If K_1 is a single point, then proof is complete.
- Otherwise, by minimality the compact (because is continuous) set $K_1 K_1$ contains a point other than the origin, so exists $V \in N(0)$ such that $\bar{V} \not\supset K_1 K_1$.
- For some $|\alpha| \leq 1$, there is a convex neighborhood $V_1 \in N(0)$ such that $\alpha V_1 \subset V$.
- By equicontinuity of \mathfrak{G} on $K \supset K_1$, there is $U_1 \in N(0)$ such that for $k_1, k_2 \in K$ and $k_1 k_2 \in U_1$ we have $\mathfrak{G}(k_1 k_2) \subset V_1 >$
- Because $T \in \mathfrak{G}$ is invertible (\mathfrak{G} is a group), T maps open sets to open sets and $T(A \cap B) = TA \cap TB$ for sets A, B.
- Since T is linear, for any A

$$T\operatorname{conv}(A) = \operatorname{conv}(TA)$$
 (8)

So

$$U_2 := \operatorname{conv}(\mathfrak{G}U_1 \cap (K_1 - K_1)) = \operatorname{conv}(\mathfrak{G}(U_1 \cap (K_1 - K_1))) \subset V$$

is relatively open in $K_1 - K_1$ and satisfies $\mathfrak{G}U_2 = U_2 \not\supset K_1 - K_1$.

- By continuity, $\mathfrak{G}U_2 = \overline{\mathfrak{G}U_2}$.
- Let $\delta = \inf\{a : a > 0, aU_2 \supset K 1 K_1\} \ge 1$, by compactness $\delta < \infty$. Let $U := \delta U_2$.
- For each $\varepsilon \in (0,1)$

$$(1+\varepsilon)U\supset K_1-K_1\not\subset (1-\varepsilon)\overline{U}$$

- Because $(1-1/4n)\bar{U} \not\supset K_1-K_1$, we have $K_2 \neq K-1$.
- K_2 is closed and convex
- Further, since $T(a\bar{U}) \subset a\bar{U}$ for $T \in \mathfrak{G}$, we have

$$T(a\bar{U}+k) \subset a\bar{U}+Tk$$
 for all $T \in \mathfrak{G}, k \in K_1$

• Recalling $TK_1 \subset K_1$ for $T \in \mathfrak{G}$ and that \mathfrak{G} is a group, we find that $TK_1 = K_1\mathfrak{G}K_2 = \underline{\text{a contradic-}}$ Finishtion

Theorem 18 (Existence of Haar Measure)

{thm:haar-me asure}

 $\overline{G \text{ compact group, } \mathcal{C}(G) \text{ continuous maps } G \to \mathbb{R}}$. Then there is a unique linear form $m : \mathcal{C}(G) \to \mathbb{R}$ such that

- 1. $m(f) \ge 0$ for $f \ge 0$ (positive)
- 2. m(1) = 1 (normalized)
- 3. m(sf) = m(f) where $sf(g) = f(s^{-1}g)$ for $s, g \in G$ (left invariant)
- 4. $m(f_s) = m(f)$ where $f_s(g) = f(gs)$ (right invariant)

2 Lecture 2: Measure theory continued

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Today we will start off by completing the construction of the Haar measure. All metric spaces are Hausdorff.

Recall that Theorem 17 (Kakutani Fix Point Theorem) requires equicontinuity of a set of continuous maps. The following result gives an alternative characterization of equicontinuity.

Theorem 19 (Generalization of Arzela-Ascoli)

{thm:arzela-ascoli}

Let X be a compact topological space. A subset of function $F \subset \mathcal{C}(X)$ is relatively compact in topology induced by uniform norm $\Leftrightarrow F$ is equicontinuous and pointwise bounded.

 $f \mapsto {}_s f$ linear transform of $\mathcal{C}(G)$, $\|{}_s f - {}_s g\| = \|f - g\|$, G acts as a group of isometries on $\mathcal{C}(G)$. In particular, this group is equicontinuous.

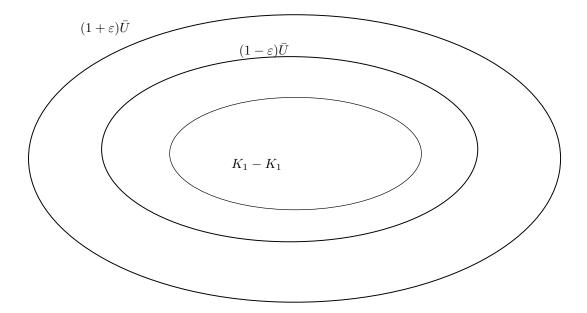


Figure 1: Sketch of proof of Kakutani's theorem

Proof of Theorem 18. Fix $f \in \mathcal{C}(G)$. Let \mathcal{C}_f denote the convex hull of all left translates of f. Elements $g \in \mathcal{C}_f$ are finite sums of form

$$g(x) = \sum_{\text{finite}} a_i f(s_i x), \qquad a_i > 0, \sum_{\text{finite}} a_i = 1$$

Clearly $||g||_{\infty} = \max\{|g(x)| : x \in G\} \le ||f||_{\infty}$, thus sets of the form $C_f(x) = \{g(x) : g \in C_f\}$ are bounded.

Since G is compact, f is uniformly continuous hence for $\varepsilon > 0$ there exists a neighborhood $V = V_{\varepsilon}$ of the identity $e \in G$ such that

$$y^{-1}x \in V \Rightarrow |f(x) - f(y)| \le \varepsilon$$

Since $(s^{-1}y)^{-1}s^{-1}x = y^{-1}x$, we also have .

By Theorem 19 (Generalization of Arzela-Ascoli), C_f relatively compact in C(G), so its closure $K_f = \overline{C_f}$ is compact convex. G acts by left translation on C(G) and leaves C_f invariant, hence K_f invariant as well. By Theorem 17 (Kakutani Fix Point Theorem), there is a fixed point $g \in K_f$ of this action of G on K_f which satisfies

$${}_sg=g\;(\forall s\in G)\quad\Rightarrow\quad g(s^{-1})={}_sg(e)=g(e)=c\;(\forall s\in G)$$

for some constant c.

By definition of K_f , for any $\varepsilon > 0$ there is $\{s_1, \ldots, s_n\} \subset G$ and $a_i > 0$ such that

$$\sum_{i=1}^{n} a_i = 1, \quad \text{and} \quad \left| c - \sum_{i=1}^{n} a_i f(s_i x) \right| < \varepsilon \qquad (\forall x \in G)$$

We first show there is only one constant function in K_f , so the fix point $g \in K_f$ is unique.

Start with same construction as before expect now use right translations of f (i.e. using the opposite group G' of G, or the function $f' = f(x^{-1})$, obtaining relatively compact set \mathcal{C}'_f with compact convex closure K'_f containing constant function c').

Opposite group

The opposite group g' of the group G is the group that coincides with G as a set but has group operation $(x,y) \mapsto y^{-1}x^{-1}$

It will be enough to show c = c', since all constants $c \in K_f$ must be equal to one chosen constant $c' \in K_f'$ and conversely.

By construction, exists finite combination of right translates close to c' i.e.

$$|c' - \sum_j b_j f(xt_j)| < \varepsilon \qquad \text{(for some } t_j \in G, \, b_j > 0 \text{ with } \sum_j b_j = 1) \tag{9} \qquad \text{(g)} \qquad \text{(general combosons)} \qquad \text{(g)} \qquad \text{(gose-to-cons)} \qquad \text{(gose-to-cons)} \qquad \text{(here)} \qquad \text{(gose-to-cons)} \qquad \text{(gose-to-cons)}$$

Multiply by a_i and put $x = s_i$ to get

$$|c'a_i - \sum_j a_i b_j f(s_i t_j)| < \varepsilon a_i$$

Summing over i

$$|c' - \sum_{i,j} a_i b_j f(s_i t_j)| = |c' \sum_i a_i - \sum_{i,j} a_i b_j f(s_i t_j)| < \varepsilon \sum_i a_i = \varepsilon$$

Operating symmetrically on Eq. (9) (mult by b_j , $x = t_j$)

finish

The following properties are obvious:

- m(1) = 1 since $K_f = \{1\}$ for f = 1
- $m(f) \ge 0$ if $f \ge 0$
- m(af) = am(f) for any $a \in \mathbb{R}$ (since $K_{af} = K_f$)
- $m(s, f) = m(f) = m(f_s)$ (by uniqueness)

It remains to show that m is additive (hence linear). Take $f, g \in \mathcal{C}(G)$ and start with Eq. (9) with c = m(f), further letting $h(x) = \sum_i a_i g(s_i x)$. Since $h \in \mathcal{C}_g$, we have $\mathcal{C}_h \subset \mathcal{C}_g$ hence $K_h \subset K_g$. But K_g contains only one constant so in fact m(h) = m(g).

We can write

$$|m(h) - \sum b_j h(t_j x)| < \varepsilon$$

for finitely many $t_i \in G$ and $b_i > 0$, $\sum_i b_i = 1$.

Using this definition of h and m(h) = m(g) gives

$$|m(g) - \sum_{i,j} a_i b_j g(s_i t_j x)| < \varepsilon$$

However, multiplying Eq. (9) by b_j and replacing x, adding (9) and (10) gives

$$|m(f) + m(g) - \sum_{i,j} a_i b_j (f+g)(s_i t_j x)| < 2\varepsilon$$

Thus $m(f) + m(g) \in K_{f+g}$, establishing additivity. Note that the only constant in K_{f+g} is .

7??

We now want to head towards some integration against probability measures.

Definition 20

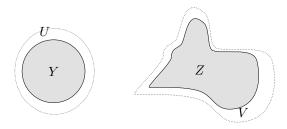
A topological space X is **normal** if for any disjoint closed sets Y and Z there exists disjoint open sets U and V such that $Y \subset U$ and $Z \subset V$.

Definition 21

X is **completely regular** if for all $y \in X$ and every closed $Z \subset X \setminus \{y\}$ there exists $f: X \to [0,1]$ continuous such that f(y) = 0 and f(z) = 1 for all zinZ.

Lemma 22 (Urysohn)

Every normal space is completely regular.



{fig:normaltopologicalspace}

Figure 2: Normal topological spaces admit separating closed sets with two disjoint open sets

Definition 23

m a set function on X with values in $[0, +\infty]$ such that $m(\emptyset) = 0$. $A \subset X$ is **Carathéodory measurable** wrt m (Carathéodory m-measurable) if, for every $E \subset X$ we have

$$m(E \cap A) + m(E \setminus A) = m(E)$$

Let \mathfrak{m}_m denote class of all Carathéodory m-measurable sets.

Theorem 24

- 1. \mathfrak{m}_m is an algebra, m is additive on \mathfrak{m}_m
- 2. For all sequences of pairwise disjoint $A_i \in \mathfrak{m}_m$: ???
- 3. If m is an outer measure on X, then \mathfrak{m}_m is a σ -algebra, m is countably additive on \mathfrak{m}_m , and m is complete on \mathfrak{m}_m

Example 25

Let \mathfrak{X} be a family of subsets of X such that $\emptyset \in \mathfrak{X}$. Given $\tau : \mathfrak{X} \to [0, +\infty]$ with $\tau(\emptyset) = 0$, set

$$\mathfrak{m}(A) = \inf \left\{ \sum_{n=1}^{\infty} \tau(X_n) : X_n \in \mathfrak{X}, A \subset \bigcup_{n=1}^{\infty} X_n \right\}$$

where $\mathfrak{m}(A) = \infty$ in the absence of such sets X_n . Then \mathfrak{m} is an outer measure, denoted τ^* .

The **Borel** σ -algebra, $\mathcal{B}(X)$, is generated by all open sets.

Let $\mathcal{B}a(x)$ denote the **Baire** σ -algebra, generated by

$$\{x \in X : f(x) > 0\}$$
 (10) {{eq:function ally-open}}

where f is a continuous function on X. It is the smallest σ -algebra where all continuous functions are measurable.

The sets of the form Eq. (10) are called *functionally open*.

Lemma 26

In a metric space X, any closed set is the set of zeros of a continuous function. Hence, $\mathcal{B}(X) = \mathcal{B}a(X)$.

Lemma 27 (Baire sets are countably determined)

Every Baire set is determined by some countable family of functions, i.e. has the form

$$\{x: (f_i(x))_i: i < \infty, f_i \in \mathcal{C}(X)\}$$

A consequence of the monotone class theorem ???

Throughout, we consider (signed) measures of bounded variation unless explicitly denoted otherwise.

Definition 28

Let X be a topological space.

- A countably additive measure on $\mathcal{B}(X)$ is called a **Borel measure**
- A countably additive measure on $\mathcal{B}a(X)$ is called a **Baire measure**
- A Borel measure μ on X is called **Radon measure** if every $B \in \mathcal{B}(X)$ can be approximated from the inside by compact sets: for $\varepsilon > 0$ exists $K_{\varepsilon} \subset B$ such that $|\mu|(B \setminus K_{\varepsilon}) < \varepsilon$.

When are two Borel measures equal?

Lemma 29

If two Borel measures coincide on all open sets, thne they coincide on all Borel sets.

Proof. Split $\mu = \mu^+ - \mu^-$ and notice that each of the two components are nonnegative and coincide on open sets. By monotone class theorem, $\mu^+ = \mu^-$.

 μ is Radon iff $|\mu|$ is Radon iff both μ^+ and μ^- are Radon.

Inner and outer approximatino of measures on \mathbb{R}^n :

Theorem 30

 $\mu \geq 0$ on $\mathcal{B}(\mathbb{R}^n)$, then any Borel set $B \subset \mathbb{R}^n$ and any $\varepsilon > 0$ exists U_{ε} open and F_{ε} closed such that $F_{\varepsilon} \subset A \subset U_{\varepsilon}$ and $\mu(U_{\varepsilon} \setminus F_{\varepsilon}) < \varepsilon$.

Proof. Let \mathcal{A} the class of all sets $Ain\mathcal{B}$ such that $F_{\varepsilon} \subset A \subset U_e$ and $\mu(U_{\varepsilon} \setminus F_{\varepsilon}) < \varepsilon$ for some closed set F_{ε} and open set U_{ε} .

Every closed A is in \mathcal{A} , since we can take $F_{\varepsilon} = A$ and U_{ε} some open δ -nbd and consider $\delta \to 0$.

It suffices to show that \mathcal{A} is a σ -algebra, since the closed sets generate \mathcal{B} . \mathcal{A} is closed wrt complements, so it remains to verify closure under countable union.

Let $A_j \in \mathcal{A}$, $\varepsilon > 0$. Then exists closed F_j and open U - j such that $F_j \subset A_j \subset U - j$ and $\mu(U_j \setminus F_j) < \varepsilon 2^{-j}$ for $j \in \mathbb{N}$.

The set $U = U_{i=1}^{\infty} U_j$ is open, and $Z_k = U_{i=1}^k F_j$ is closed.

Observe $Z_k \subset \bigcup_{j=1}^{\infty} A_j \subset U$ and for sufficiently large $k \ \mu(U \setminus Z_k) < \varepsilon$.

Indeed, $\mu(\cup_{i=1}^{\infty}\mu_{i}\setminus F_{i})<2$

□ - (?'

finish

Definition 31

Set function $\mu \geq 0$ defined on $\mathcal{A} \subset 2^X$ is **tight** on \mathcal{A} if $\forall \varepsilon > 0$ exists compact $K_{\varepsilon} \subset X$ such that $\mu(A) < \varepsilon$ for all $A \in \mathcal{A}$ that does not meet K_{ε} .

Additive set function μ of bounded variation on an algebra is **tight** if its total variation $|\mu|$ is tight.

A Borel measure is tight iff $\forall \varepsilon > 0$ exists compact K_{ε} such that $|\mu|(X \setminus K_{\varepsilon}) < \varepsilon$ (the "total variation measure").

The second definition is necessary to handle Baire sets.

Definition 32

 μ is **regular** if $\forall A \in \mathcal{A}, \varepsilon > 0$, $\exists F_{\varepsilon}$ closed such that $F_{\varepsilon} \subset A$, $A \setminus F_{\varepsilon} \in \mathcal{A}$, and

Theorem 27 implies any Borel measure on \mathbb{R}_n is regular, and the same proof works to show any Borel measure on metric space is regular.

Corollary 33 (Baire measures are regular)

Every Baire measure μ on topological space X is regular. Moreover, for every Baire set E and $\varepsilon > 0$, there exists a continuous function f on X such that $f^{-1}(0) \subset E$ and $|\mu|(E \setminus f^1(0)) < \varepsilon$.

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